NBA(National Basketball Association)



Understanding Data

Position Of Player

PG-- Point guard **SF**-- Small forward **SG**-- Shooting guard **PF**-- Power forward **C**-- Center

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [3]: #Reading the CSV files
df=pd.read_csv("nba.csv",encoding='unicode_escape')

In [10]: df.shape

Out[10]: (457, 8)

In [11]: df.head()

Out[11]:		Name	Team	Number	Position	Age	Weight	College	Salary
	0	Avery Bradley	Boston Celtics	0	PG	25	180	Texas	7730337.0
	1	Jae Crowder	Boston Celtics	99	SF	25	235	Marquette	6796117.0
	2	John Holland	Boston Celtics	30	SG	27	205	Boston University	NaN
	3	R.J. Hunter	Boston Celtics	28	SG	22	185	Georgia State	1148640.0
	4	Jonas Jerebko	Boston Celtics	8	PF	29	231	NaN	5000000.0

In [12]: df.tail()

Out[12]:		Name	Team	Number	Position	Age	Weight	College	Salary
	452	Trey Lyles	Utah Jazz	41	PF	20	234	Kentucky	2239800.0
	453	Shelvin Mack	Utah Jazz	8	PG	26	203	Butler	2433333.0
	454	Raul Neto	Utah Jazz	25	PG	24	179	NaN	900000.0
	455	Tibor Pleiss	Utah Jazz	21	С	26	256	NaN	2900000.0
	456	Jeff Withey	Utah Jazz	24	С	26	231	Kansas	947276.0

In [14]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 457 entries, 0 to 456
Data columns (total 8 columns):
# Column
            Non-Null Count Dtype
             -----
0 Name
            457 non-null
                             object
                             object
1
    Team
             457 non-null
    Number 457 non-null
                             int64
3 Position 457 non-null
                            object
4
   Age
            457 non-null
                             int64
              457 non-null
    Weight
                             int64
6 College 373 non-null
7 Salary 446 non-null
                             object
                             float64
dtypes: float64(1), int64(3), object(4)
memory usage: 28.7+ KB
```

Data Cleaning

```
In [16]: pd.isnull(df).sum()
Out[16]: Name
                      0
         Team
                      0
         Number
                      0
         Position
                      0
         Age
                      0
         Weight
                      0
         College
                     84
         Salary
                     11
         dtype: int64
In [17]: df.dropna(inplace=True)
In [18]: df.drop_duplicates(inplace=True)
In [19]: pd.isnull(df).sum()
Out[19]: Name
         Team
                     0
         Number
                     0
         Position
                     0
         Age
                     0
         Weight
                     0
         College
                     0
         Salary
                     0
         dtype: int64
```

Data Transformation

Creating a new column (BMI) using a fixed Height Value

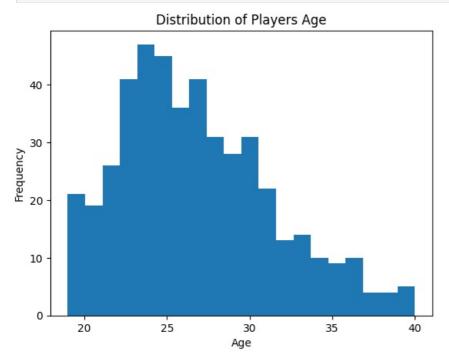
Out[26]:		Name	Team	Number	Position	Age	Weight	College	Salary	ВМІ
	0	Avery Bradley	Boston Celtics	0	PG	25	180	Texas	7730337.0	22.496000
	1	Jae Crowder	Boston Celtics	99	SF	25	235	Marquette	6796117.0	29.369778
	2	John Holland	Boston Celtics	30	SG	27	205	Boston University	NaN	25.620444
	3	R.J. Hunter	Boston Celtics	28	SG	22	185	Georgia State	1148640.0	23.120889
	4	Jonas Jerebko	Boston Celtics	8	PF	29	231	NaN	5000000.0	28.869867

EXPLORATORY DATA ANALYSIS

```
In [29]: df[['Age','Weight','BMI']].describe()
Out[29]:
                                Weight
                       Age
          count 457.000000 457.000000 457.000000
                  26.938731 221.522976
                                         27.685449
          mean
                             26.368343
                                          3.295457
            std
                   4.404016
            min
                  19.000000 161.000000
                                         20.121422
            25%
                  24.000000 200.000000
                                         24.995556
                                         27.495111
            50%
                  26.000000 220.000000
            75%
                  30.000000 240.000000
                                         29.994667
                  40.000000 307.000000
                                         38.368178
```

From above chart we can get information the Mean of Age and BMI is 27 and 28 approx respectively

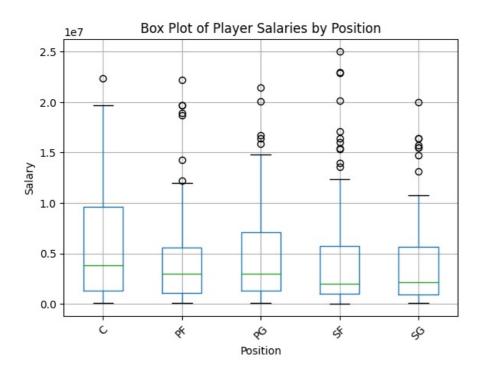
```
In [34]: plt.hist(df['Age'],bins=20)
    plt.xlabel('Age')
    plt.ylabel('Frequency')
    plt.title('Distribution of Players Age')
    plt.show()
```



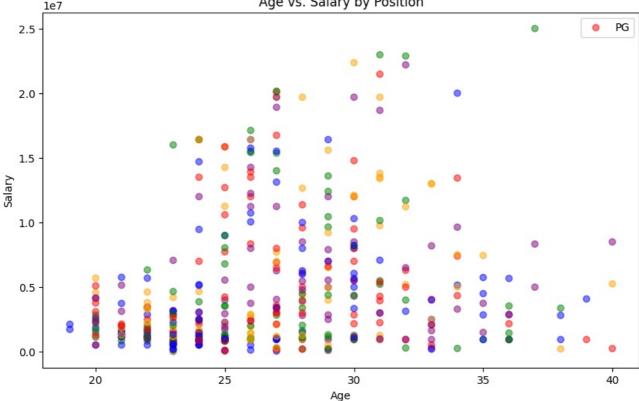
From the above histogram it shows distributions of Players age with intervals(20)

```
In [35]: plt.figure(figsize=(10, 6))
    df.boxplot(column='Salary', by='Position')
    plt.ylabel('Salary')
    plt.title('Box Plot of Player Salaries by Position')
    plt.suptitle('')
    plt.xticks(rotation=45)
    plt.show()
```

<Figure size 1000x600 with 0 Axes>



```
In [36]: # Scatter plot of 'age' vs. 'salary' by position
  plt.figure(figsize=(10, 6))
  colors = {'PG': 'red', 'SG': 'blue', 'SF': 'green', 'PF': 'purple', 'C': 'orange'}
  plt.scatter(df['Age'], df['Salary'], c=df['Position'].map(colors), alpha=0.5)
  plt.xlabel('Age')
  plt.ylabel('Salary')
  plt.title('Age vs. Salary by Position')
  plt.legend(colors)
  plt.show()
```



```
In [5]: #Top Players in terms of salary
        top players=df.nlargest(10, 'Salary')
        print(top_players)
                                                       Number Position
                                                                               Weight \
                        Name
                                                 Team
                                                                          Age
       109
                 Kobe Bryant
                                  Los Angeles Lakers
                                                            24
                                                                     SF
                                                                           37
                                                                                  212
       169
                LeBron James
                                 Cleveland Cavaliers
                                                                           31
                                                                                  250
                                                            23
                                                                     SF
       33
             Carmelo Anthony
                                     New York Knicks
                                                                      SF
                                                                           32
                                                                                  240
       251
               Dwight Howard
                                     Houston Rockets
                                                                      C
                                                                           30
                                                                                  265
                                                            12
       339
                  Chris Bosh
                                          Miami Heat
                                                             1
                                                                     PF
                                                                           32
                                                                                  235
       100
                  Chris Paul
                                                                     PG
                                                                                  175
                                Los Angeles Clippers
                                                             3
                                                                           31
       414
                Kevin Durant
                               Oklahoma City Thunder
                                                            35
                                                                     SF
                                                                           27
                                                                                  240
       164
                Derrick Rose
                                                                     PG
                                                                           27
                                                                                  190
                                       Chicago Bulls
                                                             1
       349
                 Dwyane Wade
                                           Miami Heat
                                                             3
                                                                     \mathsf{SG}
                                                                           34
                                                                                  220
       23
                 Brook Lopez
                                       Brooklyn Nets
                                                            11
                                                                      C
                                                                           28
                                                                                  275
                  College
                                Salary
       109
                      NaN
                           25000000.0
       169
                           22970500.0
                      NaN
       33
                 Syracuse
                           22875000.0
       251
                      NaN
                           22359364.0
       339
             Georgia Tech
                           22192730.0
       100
             Wake Forest
                           21468695.0
       414
                    Texas
                           20158622.0
       164
                  Memphis
                           20093064.0
       349
                Marquette
                           20000000.0
       23
                 Stanford
                           19689000.0
```

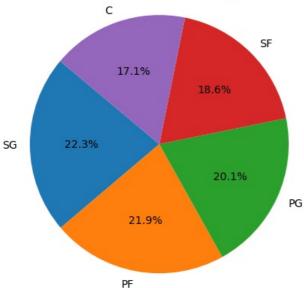
The above data shows the top-10 players having highest salary

```
In [8]: #Top colleges showing participation strength
        top_college=df['College'].value_counts().nlargest(5)
        print(top college)
       College
       Kentucky
                          22
                          20
       Duke
       Kansas
                          18
                          16
       North Carolina
       UCLA
                          15
       Name: count, dtype: int64
```

From above analysis we can view the Top college in participation is Kentucky

```
In [15]: #Position Distribution using charts
         position_counts = df['Position'].value_counts()
         plt.pie(position_counts, labels=position_counts.index, autopct='%1.1f%%', startangle=140)
         plt.title('Position Distribution of Players')
         plt.axis('equal')
         plt.show()
```

Position Distribution of Players



plt.ylabel('Average Salary')

plt.xticks(rotation=45)

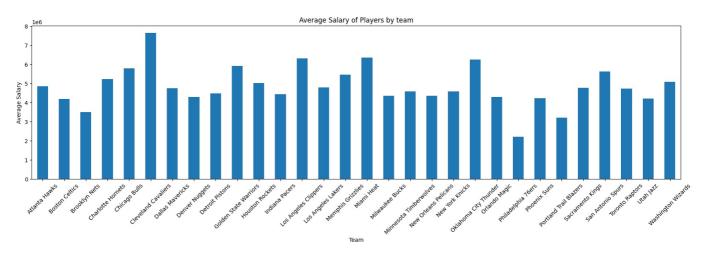
plt.show()

plt.title('Average Salary of Players by team')

Above Pie charts gives the visuals that the top position distribution of players goes to SG followed by PF,PG,SF,C

```
In [16]: #Team Analysis:
           avg salary by team=df.groupby('Team')['Salary'].mean()
           print(avg_salary_by_team)
          Team
          Atlanta Hawks
                                         4.860197e+06
         Boston Celtics
                                         4.181505e+06
                                       4.1015052
3.501898e+06
          Brooklyn Nets
                                       5.222728e+06
          Charlotte Hornets
          Chicago Bulls
                                       5.785559e+06
         Cleveland Cavaliers 7.642049e+06
Dallas Mavericks 4.746582e+06
          Denver Nuggets
                                        4.294424e+06
         Detroit Pistons 4.477884e+06
Golden State Warriors 5.924600e+06
Houston Rockets 5.018868e+06
         Golden State no.
Houston Rockets
5.01000001
4.450122e+06
         Los Angeles Clippers 6.323643e+06
Los Angeles Lakers 4.784695e+06
Memphis Grizzlies 5.467920e+06
          Miami Heat
                                       6.347359e+06
         Milwaukee Bucks 4.350220e+06
Minnesota Timberwolves 4.593054e+06
New Orleans Pelicans 4.355304e+06
         Minnesota Illinuerinos...

New Orleans Pelicans
4.355304e+vo
4.581494e+06
          Oklahoma City Thunder 6.251020e+06
         Orlando Magic
                                    4.2372.2
2.213778e+06
                                         4.297248e+06
          Philadelphia 76ers
                                        4.229676e+06
          Phoenix Suns
         Portland Trail Blazers 3.220121e+06
         Sacramento Kings 4.//objects 5.629516e+06
          Toronto Raptors
                                        4.741174e+06
                                         4.204006e+06
          Utah Jazz
          Washington Wizards
                                          5.088576e+06
          Name: Salary, dtype: float64
In [25]: plt.figure(figsize=(21,5))
           avg_salary_by_team.plot(kind='bar')
           plt.xlabel('Team')
```



Above chart shows visuals of Team(Cleveland Cavaliers) has highest salary among all teams

Conclusion

- From above data analysis we can conclude some interesting factors like Highest salary(Cleveland Cavaliers),most persons having position in SG
- Top player having highest salary is Kobe Bryant*

In []:

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